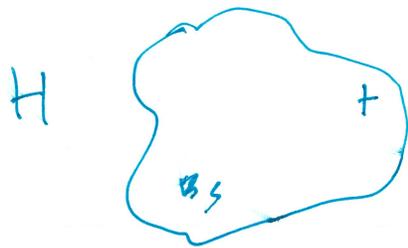


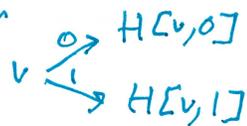
Approximating random walks with no randomness



$H = (V, E)$ $|V| = n$
 $L \leq n$ walk length

$e = \lg L$

H is two-out-regular



Goal: estimate $\Pr[H[s, X] = t] \pm \frac{1}{L}$
 $X \sim U_L$

- 1) deterministically
- 2) using $o(n)$ workspace

In fact, WalkEst $\in \mathbb{R}^L \rightarrow$ randomized logsp

Today: WalkEst $\in \mathbb{L}^2 \rightarrow$ det. $\log^2 n$ space
 via "INW PRG"

$G: \{0, 1\}^{\frac{L}{2}} \rightarrow \{0, 1\}^L$ s.t. for all H, s, t

$$\Pr_{\sigma \sim U_{\frac{L}{2}}}[H[s, G(\sigma)] = t] \approx_{\epsilon} \Pr[H[s, U_L] = t]$$

Note: above gives WalkEst $\in \text{Space}[\epsilon \frac{L}{2}]$ ^{additive error}

For now: Assume we're half way there \rightarrow of size n

$G': \{0, 1\}^{\frac{L}{2}} \rightarrow \{0, 1\}^{\frac{L}{2}}$ s.t. $\forall H, s, t$ \uparrow holds

How do we go from this to length- L walks?

Ideal: "doubling PRG"

$$G_2(\sigma_1, \sigma_2) = G^1(\sigma_1) G^1(\sigma_2)$$

Seed len: ~~2d~~ $2d'$

Is it a good PRG? Fix H, s, t

~~$$\mathbb{E}[\mathbb{E}[H[s, G^1(u_1) G^1(u_2)] = t]]$$~~

$$H_{st}(x) = \begin{cases} 1 & H[s, x] = t \\ 0 & \text{else} \end{cases}$$

~~$$\mathbb{E}_{r_1, r_2} [H[s, r_1 r_2]]$$~~

$$\mathbb{E}_{r_1, r_2} H_{st}(r_1 r_2) = \mathbb{E}_{r_1} \left[\underbrace{\mathbb{E}_{r_2} H_{st}(r_1 r_2)}_{v = H[s, r_1]} \right]$$

$$\mathbb{E}_{r_2} [H_{v+t}(r_2)]$$

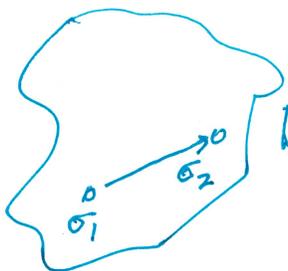
$$\approx_{\epsilon} \mathbb{E}_{r_1} \mathbb{E}_{\sigma_2} H_{st}(r_1 G^1(\sigma_2))$$

$$= \mathbb{E}_{\sigma_2} \mathbb{E}_{r_1} H_{st}(r_1 G^1(\sigma_2))$$

$$\approx_{\epsilon} \mathbb{E}_{\sigma_1, \sigma_2} H_{st}(G^1(\sigma_1) G^1(\sigma_2))$$

\Rightarrow doubling PRG G^1 is a 2ϵ -PRG for length L walks, can we do better than seed length $2d'$?

Heuristic: "choose two indep things" = "take a step on complete graph"



D on $\{0, 1\}^{d'}$

choose vtx: d' bits

choose nbr: d' bits in

Heuristic 2: complete graph \approx expander graph!

sps D is λ expander w/ deg $m \ll d'$

Then $(\sigma_1, D[\sigma_1, e_1])$ takes $d' + \log(m)$ bits to sample

$$G_2(\sigma_1, e_1) = G^1(\sigma_1) G^1(D[\sigma_1, e_1])$$

How to analyze G_2 ? Show it's a 3ε -PRG?

Idea: compare to doubling PRG

$\forall H, s, t$

$$\mathbb{E}_{\sigma_1, \sigma_2} [H_{st}(G_1(\sigma_1, \sigma_2))] \approx \mathbb{E}_{\sigma, e_1} [H_{st}(G_2(\sigma, e_1))]$$

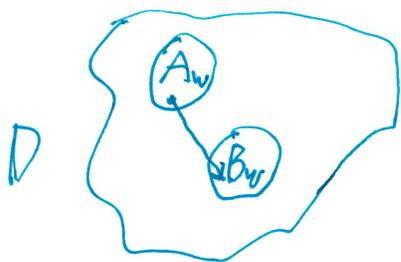
Consider midpoint of walk



$$A_w = \{ \sigma : H_{sw}(G_1(\sigma)) \}$$

$$B_w = \{ \sigma : H_{wt}(G_1(\sigma)) \}$$

$$\mathbb{E}_{\sigma_1, \sigma_2} [H_{st}(G_1(\sigma_1, \sigma_2))] = \sum_w \frac{|A_w|}{2^{d_1}} \cdot \frac{|B_w|}{2^{d_1}} = \sum_w P_w$$



Clm: $\forall w, (\sigma_1, D[\sigma_1, e_1]) \in A_w \times B_w$
with approximately correct density

$$P_{\sigma_1, e_1} [\sigma_1 \in A_w, D[\sigma_1, e_1] \in B_w] = \frac{|A_w|}{2^{d_1}} \cdot \frac{|B_w|}{m}$$

By "expander mixing lemma" $\approx \lambda P_w$

Thus G_2 ~~is~~ $n \cdot \lambda$ -approx G_1 which 2ε -approx U_L

$\Rightarrow G_2$ $(n\lambda + 2\varepsilon)$ -approx U_L

Take $\lambda \approx 1/nL$ & recurse construction 2 times

$$\Rightarrow \text{final PRG w/ seed } \log(\deg(D_1)) + \dots + \log \deg D_e \\ = 2 \cdot \log \text{poly}(nL) = O(\log^2(n) \log(nL))$$

